## S4 Text. Empirical Bayes estimator of error rates

Assume that  $\epsilon_j$  follows a prior distribution Beta(a,b) and the total number of errors across individuals, denoted by  $r_j$ , follows a binomial distribution  $Bin(t_j, \epsilon_j)$ . Under this model we can obtain the expected values of the (weighted) empirical first and second moments  $m_1 = \sum_{j=1}^M \widetilde{\epsilon}_j t_j / \sum_{j=1}^M t_j$  and  $m_2 = \sum_{j=1}^M \widetilde{\epsilon}_j^2 t_j / \sum_{j=1}^M t_j$ , where  $m_1$  and  $m_2$  are weighted by the total number of reads  $t_j$  at locus j across all individuals. We estimate hyperparameters a and b using the method of moments, equating the empirical moments to their theoretical values. We find  $\widehat{a} = B^{-1}m_1(m_1 - m_2)$  and  $\widehat{b} = B^{-1}(1 - m_1)(m_1 - m_2)$ , where  $B = m_2 - m_1 + m_1(1 - m_1)(1 - M / \sum_{j=1}^M t_j)$ .

The posterior distribution of  $\epsilon_j$  given  $r_j$  and  $t_j$  is also a beta distribution  $Beta(r_j + a, t_j - r_j + b)$ . Thus, the empirical Bayes (EB) estimator is  $E(\epsilon_j|r_j) = w_j a/(a+b) + (1-w_j)r_j/t_j$ , where  $w_j = (a+b)/(a+b+t_j)$ . In calculating the EB estimator, we use the values of  $\hat{a}$  and  $\hat{b}$  obtained by the method of moments.